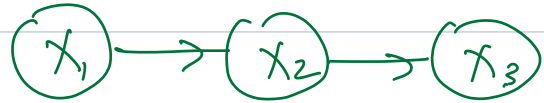


Multiple factorizations

Generative model:



i.e., $X_1 \perp\!\!\!\perp X_3 \mid X_2$.

True DAG.

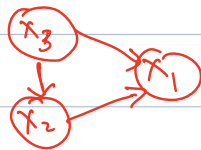
$$\text{let } p(x_1, x_2, x_3) = p(x_1) p(x_2 \mid x_1) p(x_3 \mid x_2)$$

be the true underlying distribution.

Given $p(x_1, x_2, x_3)$, we can have an alternate factorization, using chain rule:

$$p(x_1, x_2, x_3) = q(x_3) q(x_2 \mid x_3) q(x_1 \mid x_3, x_2)$$

↪ ~~*~~



Alternate model using chain rule

Since we know that $X_1 \perp\!\!\!\perp X_3 \mid X_2$, we know that there exists functions s.t.

$$p(x_1, x_2, x_3) = \phi_{32}(x_3, x_2) \phi_{12}(x_1, x_2).$$

Comparing the factorizations, this indicates that the above alternate factorization (*) should satisfy

$$q(x_1 | x_3, x_2) = q(x_1 | x_2), \text{ i.e., the}$$

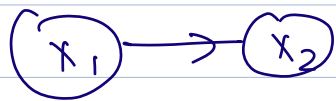
alternate graph should have a '0 weight' edge:



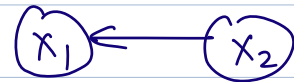
Thus, in this example, the true graph and the alternate graph has the same skeleton (i.e., undirected edges).

Question: In general, can we use CI relationships to uniquely determine the correct DAG (i.e., correct directions for edges)?

No! Counter-example: Rain and Wetness from Notes 1.



vs.



$$p(x_1, x_2) = p(x_1) p(x_2 | x_1) \quad \text{vs.} \quad p(x_1, x_2) = q(x_2) q(x_1 | x_2)$$

We cannot distinguish between these two factorizations from observational data. This again motivates the study of interventions to determine structure (here, causal relationships).